Assesment Report

on

"Customer Segmentation in E-commerce"

submitted as partial fulfillment for the award of

BACHELOR OF TECHNOLOGY DEGREE

SESSION 2024-25

in

CSE(AIML)

By

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Introduction

In the fast-evolving e-commerce industry, understanding customer behavior is crucial for sustaining competitive advantage and driving business growth.
Customer Segmentation in E-commerce: Identifying customer clusters based on purchasing habits and browsing behavior forms the core of this project.

By strategically analyzing customer purchasing patterns and browsing activities, businesses can personalize marketing efforts, optimize customer retention, and significantly enhance profitability.
This project aims to **accurately** segment customers into meaningful groups by using real-world e-commerce transaction data — ensuring that the clustering captures behavior patterns with a focus on **maximizing segmentation accuracy** through appropriate data preprocessing, feature scaling, and model selection techniques.

Through this segmentation, organizations can develop highly targeted strategies that cater to specific customer needs, thereby improving both customer satisfaction and operational efficiency.

Methodology

1. **Data Upload:**

   The provided dataset containing transaction details like InvoiceNo, StockCode, Description, Quantity, InvoiceDate , UnitPrice , CustomerID, and Country was uploaded using Google Colab's files.upload() function.

2. **Data Preprocessing:**

   - Removed missing CustomerID values.
   - Filtered out transactions with negative quantities (considered as returns).
   - Created a new feature TotalPrice = Quantity × UnitPrice.

3. **Feature Aggregation:**

   Customer-level features were generated:

   - Number of Orders
   - Total Quantity Purchased
   - Total Amount Spent

4. **Feature Scaling:**

   StandardScaler was applied to normalize the numerical features for better clustering results.

5. **Cluster Identification:**

   Used the Elbow Method to determine the optimal number of clusters. K-Means Clustering was then applied to segment the customers into meaningful groups.

6. **Visualization:**

   Visualized customer clusters using Seaborn scatter plots, highlighting patterns based on total spending and quantity purchased.

CODE:

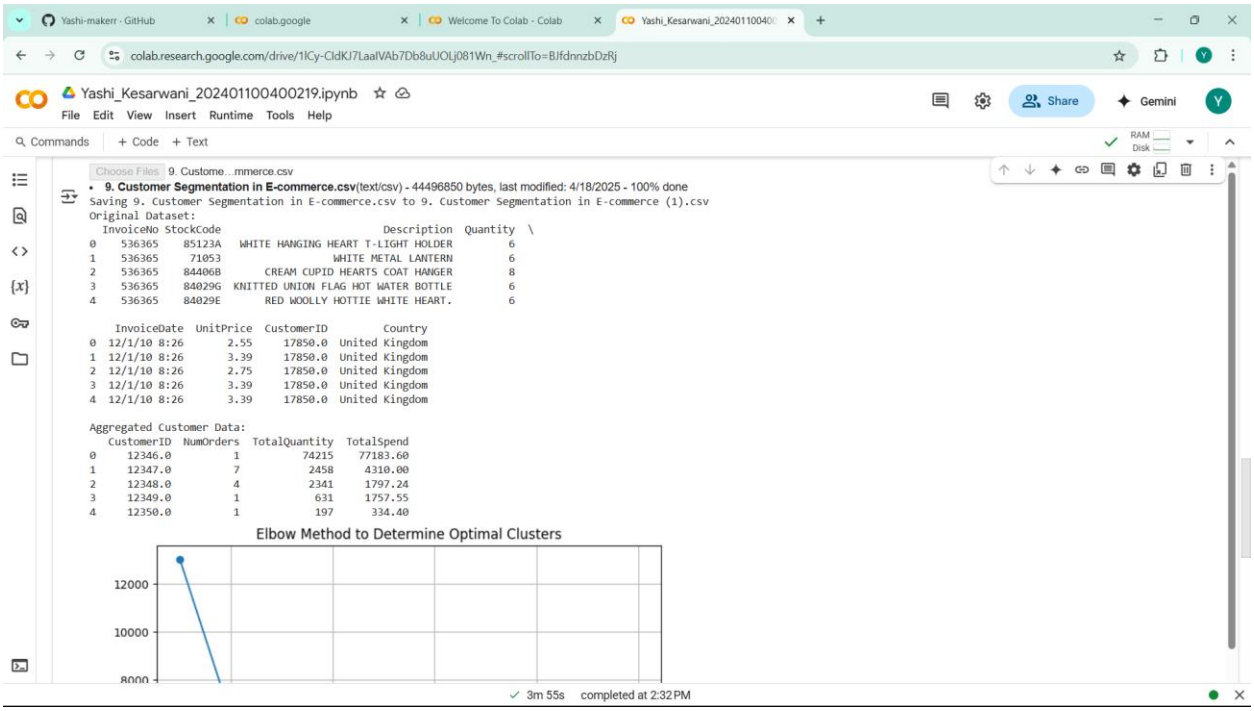
```
# 1. Import libraries

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

from sklearn.cluster import KMeans

from sklearn.preprocessing import StandardScaler


# 2. Upload CSV file (Colab / Jupyter style)

from google.colab import files

import io


# Upload the file

uploaded = files.upload()


# Read the uploaded CSV

for file_name in uploaded.keys():

    df = pd.read_csv(io.BytesIO(uploaded[file_name]), encoding='ISO-8859-1')


# 3. Display the data

print("Original Dataset:")

print(df.head())


# 4. Data Cleaning
```

```
# Remove missing CustomerIDs
df = df.dropna(subset=['CustomerID'])

# Remove negative quantities (which are returns)
df = df[df['Quantity'] > 0]

# 5. Feature Engineering
# Create TotalPrice column
df['TotalPrice'] = df['Quantity'] * df['UnitPrice']

# Group by CustomerID to get aggregated customer-level data
customer_data = df.groupby('CustomerID').agg({
    'InvoiceNo': 'nunique',     # Number of orders
    'Quantity': 'sum',          # Total quantity purchased
    'TotalPrice': 'sum'         # Total amount spent
}).reset_index()

# Rename columns for clarity
customer_data.rename(columns={
    'InvoiceNo': 'NumOrders',
    'Quantity': 'TotalQuantity',
    'TotalPrice': 'TotalSpend'
}, inplace=True)

print("\nAggregated Customer Data:")
```

```
print(customer_data.head())


# 6. Feature Scaling
scaler = StandardScaler()
X_scaled = scaler.fit_transform(customer_data[['NumOrders', 'TotalQuantity', 'TotalSpend']])


# 7. Finding the optimal number of clusters using the Elbow Method
wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, random_state=42)
    kmeans.fit(X_scaled)
    wcss.append(kmeans.inertia_)


# Plot the Elbow Curve
plt.figure(figsize=(8, 5))
plt.plot(range(1, 11), wcss, marker='o')
plt.title('Elbow Method to Determine Optimal Clusters')
plt.xlabel('Number of Clusters')
plt.ylabel('WCSS')
plt.grid()
plt.show()


# 8. Apply KMeans Clustering
# Let's assume we choose 4 clusters based on the elbow curve
kmeans = KMeans(n_clusters=4, random_state=42)
```

```
customer_data['Cluster'] = kmeans.fit_predict(X_scaled)


# 9. Final segmented customer data
print("\nCustomer Segments:")
print(customer_data.head())


# 10. Visualize the Clusters
plt.figure(figsize=(10, 6))
sns.scatterplot(
    data=customer_data,
    x='TotalSpend',
    y='TotalQuantity',
    hue='Cluster',
    palette='Set2',
    s=100
)
plt.title('Customer Segmentation based on Spend and Quantity')
plt.xlabel('Total Spend')
plt.ylabel('Total Quantity Purchased')
plt.legend(title='Cluster')
plt.grid()
plt.show()
# 11. (Optional) Save the Segmented Customers to a new CSV
customer_data.to_csv('segmented_customers.csv', index=False)
print("\nSegmented customer data saved as 'segmented_customers.csv'.")
```

OUTPUTS


Yashi_Kesarwani_20240110040219.ipynb
File Edit View Insert Runtime Tools Help
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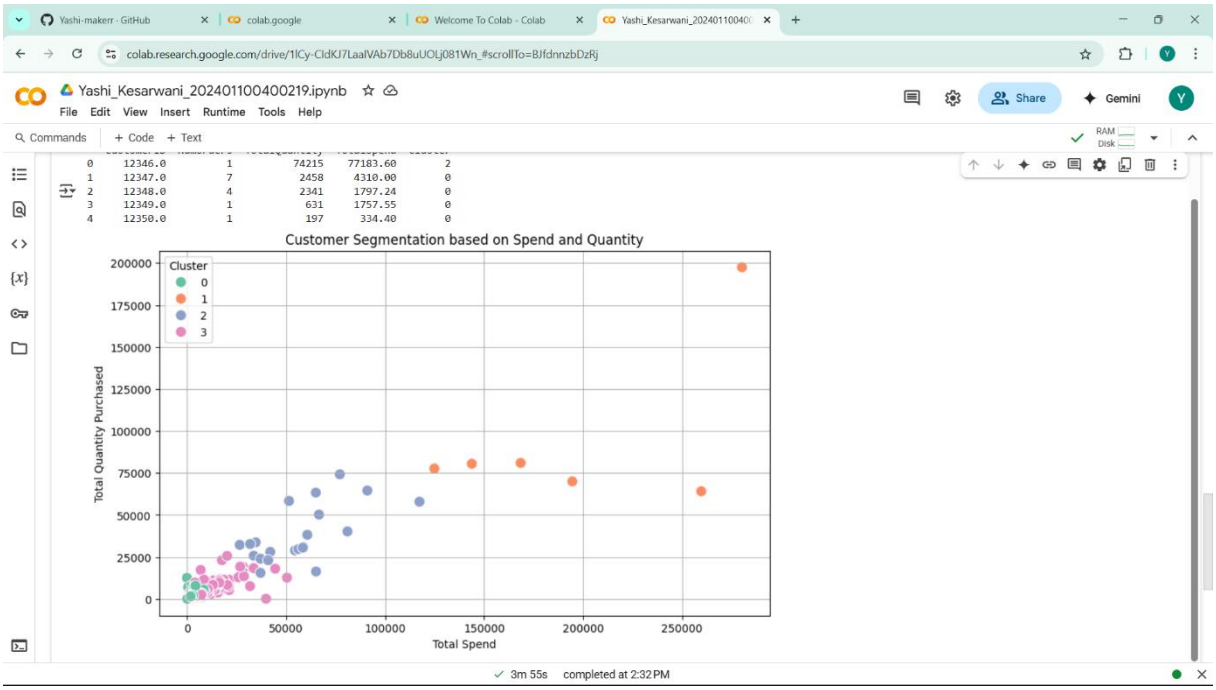
Choose Files 9. Custome...mmerce.csv
9. Customer Segmentation in E-commerce.csv(text/csv) - 44496850 bytes, last modified: 4/18/2025 - 100% done
Saving 9. Customer Segmentation in E-commerce.csv to 9. Customer Segmentation in E-commerce (1).csv
Original Dataset:
InvoiceNo StockCode Description Quantity \
0 536365 85123A WHITE HANGING HEART T-LIGHT HOLDER 6
1 536365 71053 WHITE METAL LANTERN 6
2 536365 84406B CREAM CUPID HEARTS COAT HANGER 8
3 536365 84029G KNITTED UNION FLAG HOT WATER BOTTLE 6
4 536365 84029E RED WOOLLY HOTTIE WHITE HEART. 6
InvoiceDate UnitPrice CustomerID Country
0 12/1/10 8:26 2.55 17850.0 United Kingdom
1 12/1/10 8:26 3.39 17850.0 United Kingdom
2 12/1/10 8:26 2.75 17850.0 United Kingdom
3 12/1/10 8:26 3.39 17850.0 United Kingdom
4 12/1/10 8:26 3.39 17850.0 United Kingdom
Aggregated Customer Data:
CustomerID NumOrders TotalQuantity TotalSpend
0 12346.0 1 74215 77183.60
1 12347.0 7 2458 4310.00
2 12348.0 4 2341 1797.24
3 12349.0 1 631 1757.55
4 12350.0 1 197 334.40
Elbow Method to Determine Optimal Clusters
12000
10000
8000
3m 55s completed at 2:32 PM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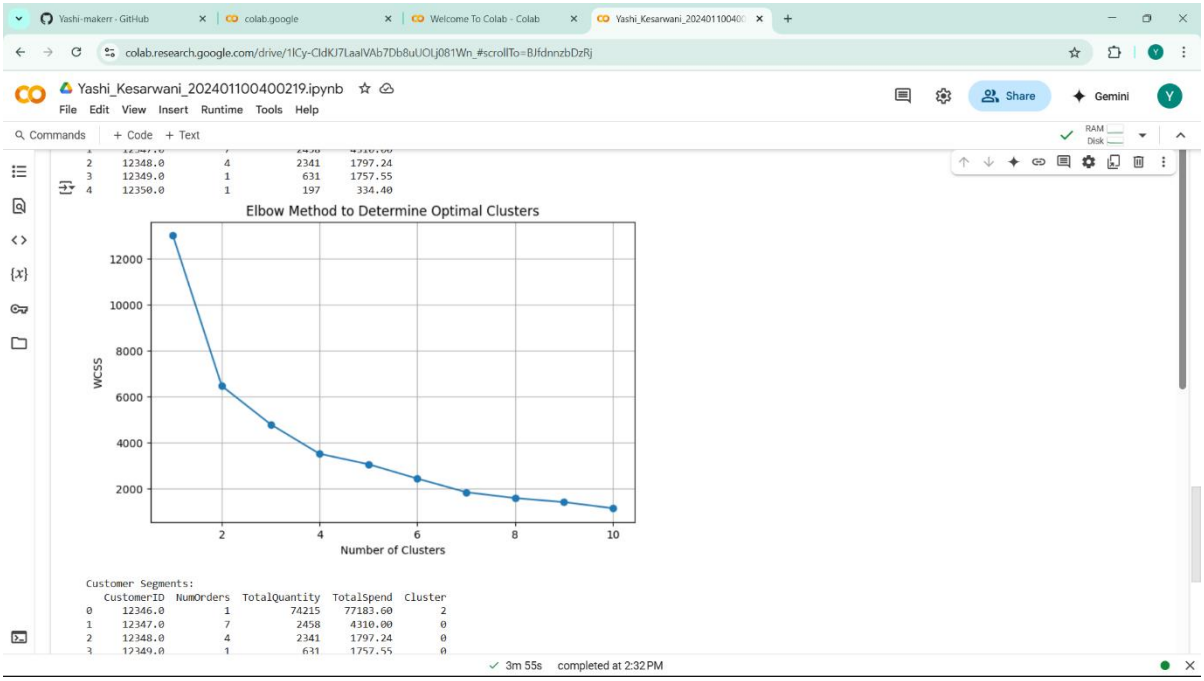
0 12346.0 1 74215 77183.60 2
1 12347.0 7 2458 4310.00 0
2 12348.0 4 2341 1797.24 0
3 12349.0 1 631 1757.55 0
4 12350.0 1 197 334.40 0
Customer Segmentation based on Spend and Quantity
Cluster
0
1
2
3
Total Quantity Purchased
Total Spend
200000
175000
150000
125000
100000
75000
50000
25000
0
0
50000
100000
150000
200000
250000
3m 55s completed at 2:32 PM


Yashi_Kesarwani_202401100400219.ipynb
File Edit View Insert Runtime Tools Help
Share
Gemini
Commands
+ Code
+ Text
2 12348.0 4 2341 1797.24
3 12349.0 1 631 1757.55
4 12350.0 1 197 334.40
Elbow Method to Determine Optimal Clusters
WCSS
Number of Clusters
12000
10000
8000
6000
4000
2000
2
4
6
8
10
Customer Segments:
CustomerID NumOrders TotalQuantity TotalSpend Cluster
0 12346.0 1 74215 77183.60 2
1 12347.0 7 2458 4310.00 0
2 12348.0 4 2341 1797.24 0
3 12349.0 1 631 1757.55 0
3m 55s completed at 2:32 PM

References/Credits

- Dataset: E-commerce Transaction Dataset (provided during the exam).
- Libraries used:
  - **Pandas** for data manipulation
  - **NumPy** for numerical operations
  - **Matplotlib** and **Seaborn** for visualization
  - **Scikit-learn** for clustering (KMeans) and scaling (StandardScaler)
- Google Colab for running the code.
- CODE : CHATGPT